

CLAIMS

What is claimed is:

1. A device for filtering electrical signals, having a number of inputs arranged spatially at a distance from one another and to supply respective pluralities of input signal samples, and a device output, to supply a plurality of filtered signal samples, the device comprising:

a number of signal processing channels, each signal processing channel being formed by a neuro-fuzzy filter to receive a respective plurality of input signal samples and to generate a respective plurality of reconstructed samples; and

an adder unit to receive said plurality of reconstructed samples and having an output to supply said plurality of filtered signal samples.

2. The device according to claim 1, further comprising routing means connected to said output of said adder unit and controlled so as first to supply said filtered signal samples back to said signal processing channels, then to supply said filtered signal samples to said device output.

3. The device according to claim 2 wherein each signal processing channel includes:

a sample input to receive alternately said input signal samples and said filtered signal samples and to supply samples of signal to be filtered;

signal feature computing unit to receive a respective plurality of samples to be filtered and to generate signal features;

a neuro-fuzzy network to receive said signal features and to generate reconstruction weights; and

a signal reconstruction unit to receive said samples to be filtered and said reconstruction weights and to generate said reconstructed samples from said samples to be filtered and said reconstruction weights.

4. The device according to claim 3 wherein said signal feature computing unit generates, for each said sample to be filtered:

- a first signal feature correlated with a position of a sample to be filtered within an operative sample window;
- a second signal feature correlated to a difference between said sample to be filtered and a central sample within said operative sample window; and
- a third signal feature correlated to a difference between said sample to be filtered and an average sample value within said operative sample window.

5. The device according to claim 1, further comprising a current-weights memory connected to said neuro-fuzzy filters and to store filter weights.

6. The device according to claim 5, further comprising a weight training unit to calculate in real time said filtering weights.

7. The device according to claim 6 wherein said weight training unit comprises:

- a training signal supply unit to supply a training signal having a known noise component;
- a weight supply unit to supply training weights;
- a spatial filtering unit to receive said training signal and said training weights and to output a filtered training signal;
- a processing unit to process said training signal and said filtered training signal and to generate a fitness value; and
- a control unit to repeatedly control said weight training unit and repeatedly receive said fitness value, said control unit being coupled to store the training weights having best fitness value in said current-weights memory.

8. The device according to claim 7 wherein said training signal supply unit includes a noise sample memory to store a plurality of noise samples, and a number of adders, one for each input of said device, each adder being coupled to receive a respective plurality of input signal samples and said noise samples, and to output a respective plurality of training signals.

9. The device according to claim 8, further comprising a switching unit having a number of changeover switch elements, one for each signal processing channel, each changeover switch element having a first input connected to a respective input of the device, a second input connected to the output of a respective adder, and an output connected to a respective signal processing channel.

10. The device according to claim 7 wherein said weight supply unit comprises a random number generator.

11. The device according to claim 8 wherein said processing unit comprises means for computing a fitness degree correlated to the signal-to-noise ratio between said filtered training signal and said noise samples.

12. The device according to claim 7, further comprising a best-fitness memory to store a best-fitness value and a best-weights value, wherein said control unit comprises comparison means for comparing said fitness value supplied by said processing unit and said best-fitness value, and writing means for writing said best-fitness memory with said fitness value, and said best-weight memory with corresponding training weights, in case said fitness value supplied by said processing unit is better than said best-fitness value.

13. The device according to claim 5, further comprising an acoustic scenario change recognition unit to receive said filtered signal samples.

14. The device according to claim 13 wherein said acoustic scenario change recognition unit includes:

a subband-splitting block to receive said filtered signal samples from said device output and to generate a plurality of sets of samples;
a features extraction unit to calculate features of each set of samples;
a neuro-fuzzy network to generate acoustically weighted samples; and
a scenario change decision unit to receive said acoustically weighted samples and to output an activation signal for activation of said weight training unit.

15. The device according to claim 14 wherein said subband splitting block includes a plurality of splitting stages in cascade.

16. The device according to claim 15 wherein each said splitting stage includes:

a first and a second filter, in quadrature to each other, to receive a stream of samples to be split and to generate each a respective stream of split samples; and
a first and a second downampler unit, each to receive a respective said stream of split samples.

17. The device according to claim 16 wherein said first filter of said splitting stages is a lowpass filter, and said second filter of said splitting stages is a highpass filter.

18. The device according to claim 14 wherein said feature extraction unit calculates energy of each set of samples.

19. The device according to claim 14 wherein said neuro-fuzzy network comprises:

fuzzification neurons to receive said signal features, and to generate first-layer outputs that define a confidence level of said signal features with respect to membership functions of a triangular type;

fuzzy AND neurons to receive said first-layer outputs and to generate second-layer outputs derived from fuzzy rules; and

a defuzzification neuron to receive said second-layer outputs and to generate an acoustically weighted sample for each of said filtered samples, using a gravity-of-gravity criterion.

20. The device according to claim 14 wherein said scenario change decision unit generates said activation signal by digitization at least one of said acoustically weighted samples.

21. The device according to claim 19, further comprising:

a clustering training network having a first input to receive said filtered signal samples from said device output, a second input to receive said acoustically weighted samples, and an output connected to the clustering weights memory, said clustering training network including:

energy calculation means for calculating the mean energy of said filtered signal samples in a preset operative window;

gravity-of-gravity calculating means for determining centers of gravity of said membership functions according to said mean energy, said gravity-of-gravity calculating means being connected and supplying said centers of gravity to said fuzzification neurons;

random generator means for randomly generating weights for said second-layer and third-layer neurons;

fitness calculation means for calculating a fitness function from said filtered signal samples and target signal samples;

fitness comparison means for comparing said calculated fitness function with a previous stored value;

storage means for storing said fitness function, said centers of gravity and said weights, in case said calculated fitness function is better than said previous stored value; and

next-activation means for activating said energy calculation means, said gravity-of-gravity calculation means, said random generator means, said fitness comparison means, and said storage means.

22. An apparatus to calculate filter weights in real time for a device that filters electrical signals, the apparatus comprising:

a training signal supply unit to supply a training signal having a known noise component;

a weight supply unit to supply training weights;

a spatial filtering unit, coupled to the training signal supply unit and to the weight supply unit, to respectively receive said training signal and said training weights and to output a filtered training signal;

a processing unit, coupled to the training signal supply unit and to the spatial filtering unit, to respectively process said training signal and said filtered training signal and to generate a fitness value; and

a control unit coupled to the processing unit to repeatedly receive said fitness value and to control storage of training weights having a best fitness value.

23. An apparatus to detect a change in acoustic scenario, the apparatus comprising:

a subband-splitting block to receive filtered signal samples and to generate a plurality of sets of samples therefrom;

a features extraction unit coupled to the subband-splitting block to calculate features of each set of samples;

a neuro-fuzzy network coupled to the features extraction unit to generate acoustically weighted samples; and

a scenario change decision unit coupled to the neuro-fuzzy network to receive said acoustically weighted samples and to output an activation signal to activate a weight training unit.

24. A method for filtering electrical signals, comprising:

receiving a plurality of streams of signal samples to be filtered; and

generating a plurality of filtered signal samples, wherein said generating includes:

filtering each stream of signal samples to be filtered through a respective neuro-fuzzy filter to generate a plurality of streams of reconstructed samples; and

adding said plurality of streams of reconstructed samples to obtain added signal samples.

25. The method according to claim 24, further comprising:

supplying said added signal samples to said neuro-fuzzy filters; and

repeating said filtering and adding to obtain said filtered signal samples and to output said filtered signal samples.

26. The method according to claim 24, further comprising weight training including:

supplying a training signal having a known noise component;

supplying filtering weights to said neuro-fuzzy filters;

filtering said signal samples to be filtered, to obtain a training filtered signal;

calculating a current fitness value from said training filtered signal samples;
comparing said fitness value with a previous fitness value; and
storing said fitness value and said filtering weights if said current fitness value is better than said previous fitness value.

27. The method according to claim 26 wherein said supplying comprises randomly generating said filtering weights.

28. The method according to claim 27 wherein said randomly generating said filtering weights, filtering, calculating a current fitness value, comparing, and storing are repeated a preset number of times.

29. The method according to claim 26 wherein said supplying a training signal comprises adding a plurality of noise samples to said filtered signal samples.

30. The method according to claim 26, further comprising recognizing acoustic scenario changes in said filtered signal samples and activating said training step.

31. The method according to claim 30 wherein said recognizing comprises:

splitting said filtered signal samples into a plurality of subbands;
filtering said subbands through clustering neuro-fuzzy filters to obtain an acoustically weighted signal; and
activating said training step if said acoustically weighted signal has a preset value.

32. The method according to claim 31 wherein said splitting includes filtering said subbands using filters having a pass band correlated to bands that are critical for a human ear.

33. The method according to claim 30, further comprising clustering training including:

calculating the mean energy of said filtered signal samples in a preset operative window;

determining centers of gravity of membership functions of said clustering neuro-fuzzy filters according to said mean energy;

calculating a fitness function from said filtered signal samples and target signal samples;

comparing said fitness function with a previous stored value; and

storing said fitness function and said centers of gravity, should said calculated fitness function be better than said previous stored value.

34. A system for filtering electrical signals, the system comprising:

a means for receiving a plurality of streams of signal samples to be filtered; and

a means for generating a plurality of filtered signal samples, including:

a means for filtering each stream of signal samples to be filtered through a respective neuro-fuzzy network to generate a plurality of streams of reconstructed samples; and

a means for adding said plurality of streams of reconstructed samples to obtain added signal samples.

35. The system of claim 34, further comprising a means for updating filter weights used by the neuro-fuzzy network.

36. The system of claim 34, further comprising a means for detecting changes in an acoustic scenario.

37. The system of claim 36, further comprising a means for training the means for detecting changes in the acoustic scenario.